

Scientific Data Management Center

(Integrated Software Infrastructure Center – ISIC)

Arie Shoshani, LBNL

**SciDAC PI Meeting
March, 2003**

(<http://sdmcenter.lbl.gov>)

Participating Institutions



Center Director: Arie Shoshani

DOE Laboratories co-ordinating PIs:

ANL: Bill Gropp

LBNL: Arie Shoshani

LLNL: Terence Critchlow

ORNL: Thomas Potok

Universities co-ordinating PIs :

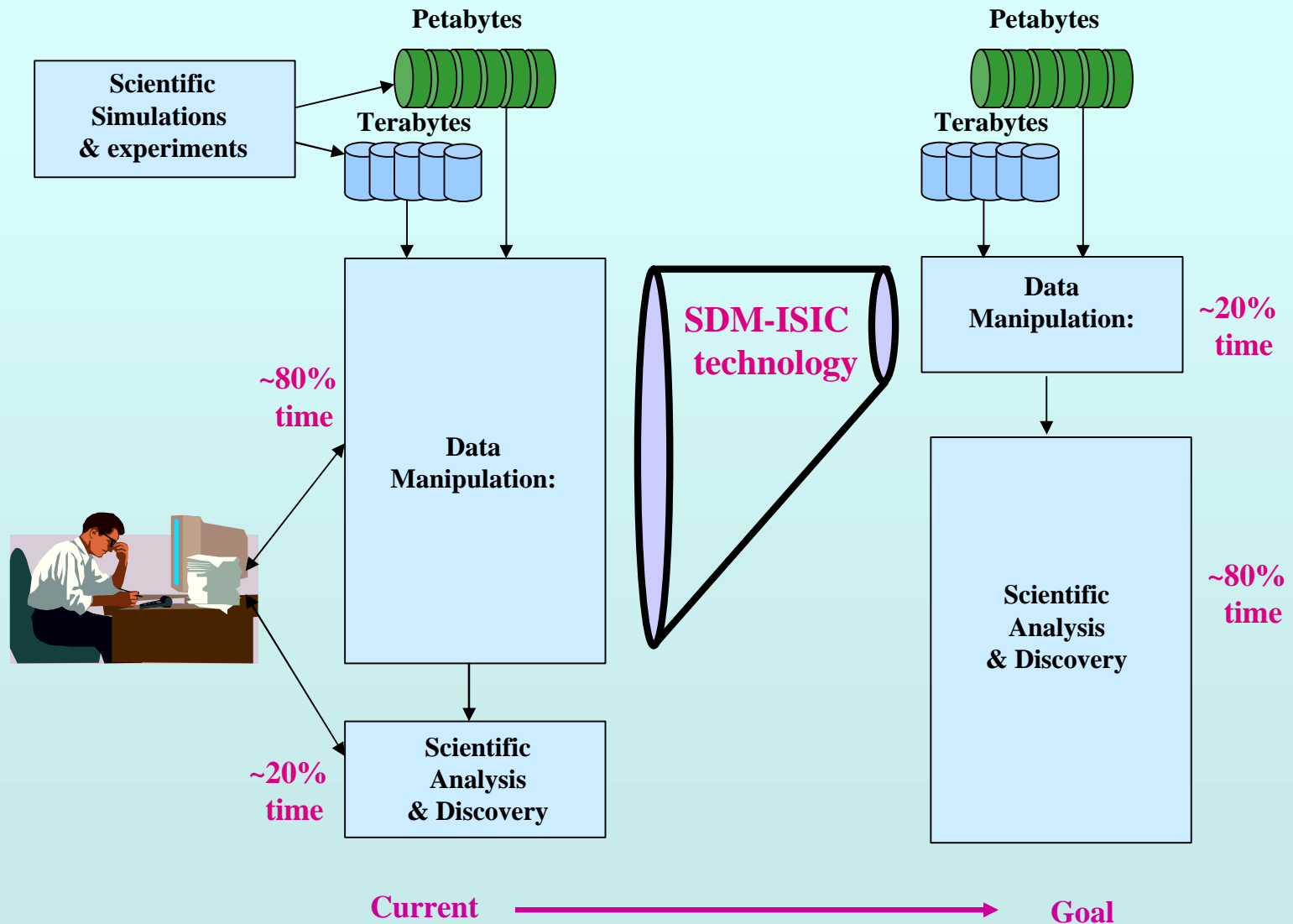
Georgia Institute of Technology: Calton Pu

North Carolina State University: Mladen Vouk

Northwestern University: Alok Choudhary

UC San Diego (Supercomputer Center): Reagan Moore

SDM Center Goal: Reduce the Data Management Overhead



Reduce the Data Management Overhead: How?



- **Efficiency**
 - Example: parallel I/O, indexing, matching storage structures to the application
- **Effectiveness**
 - Example: Access data by attributes-not files, facilitate massive data movement
- **New algorithms**
 - Example: Specialized PCA techniques to separate signals or to achieve better spatial data compression
- **Enabling ad-hoc exploration of data**
 - Example: by enabling exploratory “run and render” capability to analyze and visualize simulation output while the code is running

□ Guiding principles

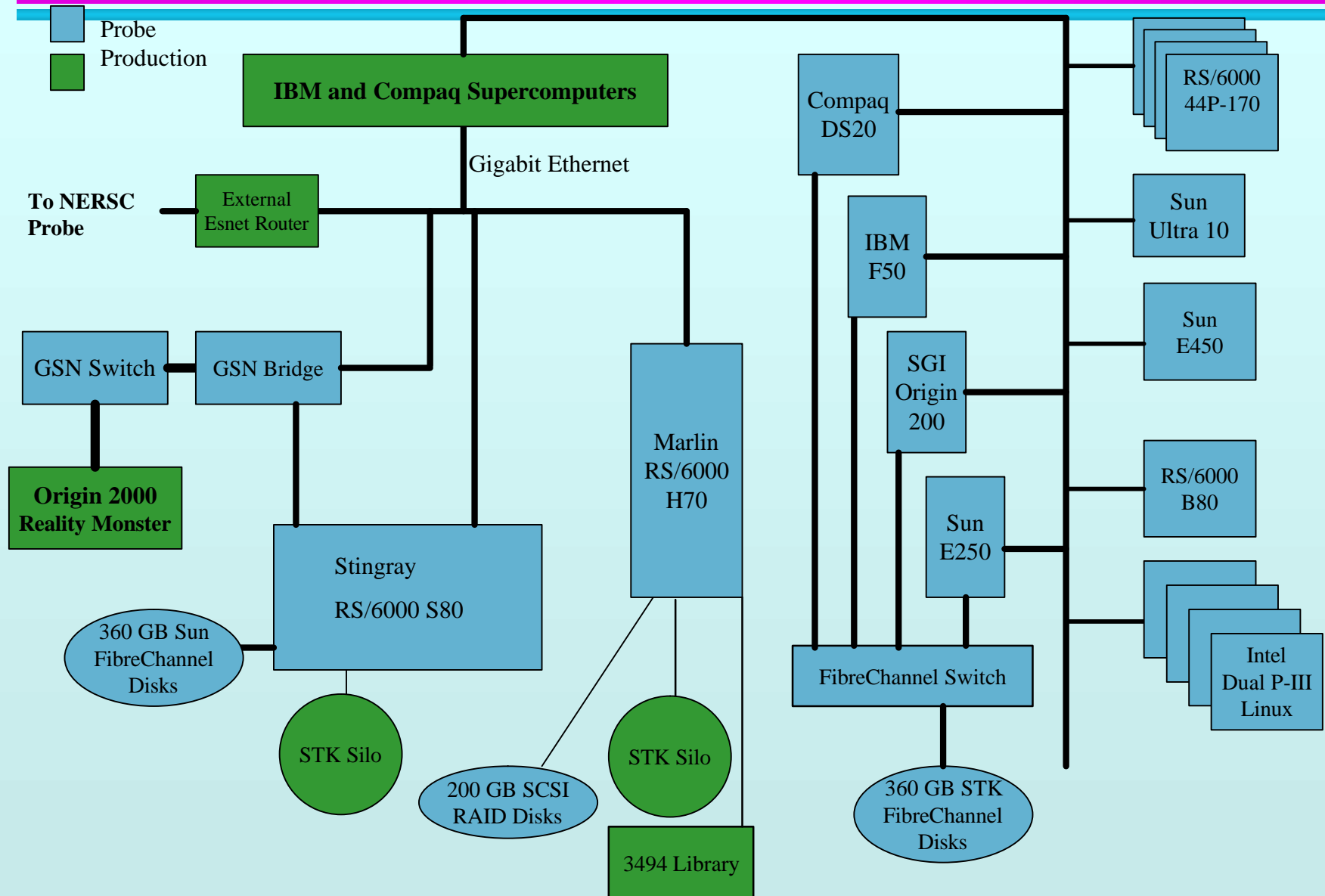
- Work with individual application scientists
- Work with specific scientific problems
- Deploy technology already developed or prototyped
- Do research/development driven by need and experience
- Re-apply techniques to new applications

□ Focus areas

- Parallel and Grid I/O Infrastructure
 - Astrophysics, Climate, Fusion
- Exploratory Analysis and Data Mining
 - Astrophysics, Climate
- Distributed, Heterogeneous Data Integration
 - Biology
- Efficient Processing and Access of Very Large Datasets
 - High Energy Physics, Combustion, Astrophysics

Test Environment: ORNL Probe – “Place to be”

Randy Burris, ORNL



Results

(Also see 5 posters)

□ Team Members

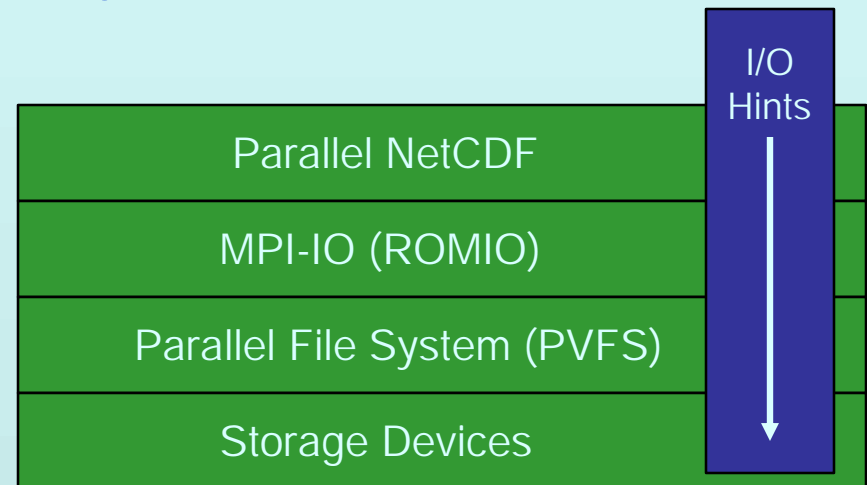
- Bill Gropp, Rob Ross, Rajeev Thakur, Rob Latham, Neill Miller (ANL)
- Alok Choudhary, Wei-Keng Liao, Jianwei Li, Avery Ching (NWU)
- Ghaleb Abdulla, Tina Eliassi-Rad (LLNL)

□ Improving software at all I/O layers

- High level interfaces
- MPI-IO
- Parallel file systems

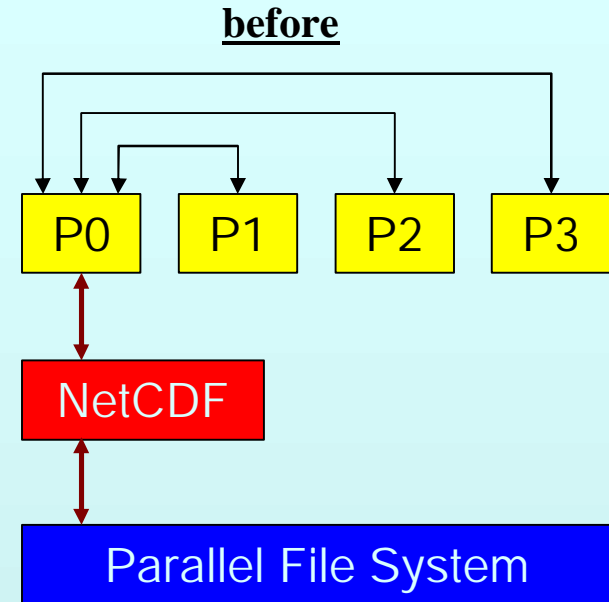
□ Enabling tighter coupling of layers

- Hints
- Structured I/O requests



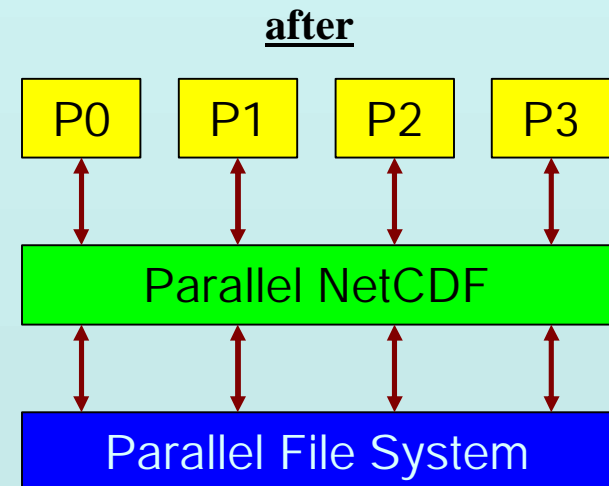
❑ Original NetCDF

- No possibility of collective optimizations
- All processes read file independently
- Writes are carried out by shipping data to a single process (sequential write)

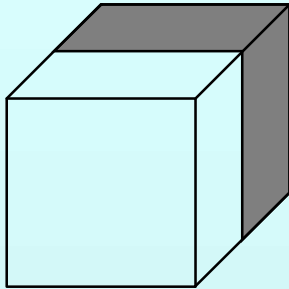
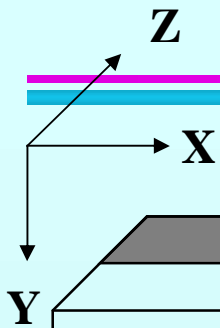


Parallel NetCDF

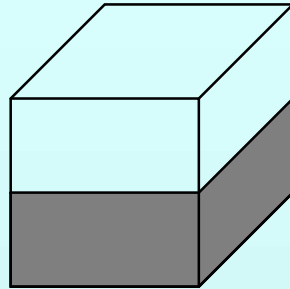
- Parallel read/write to shared NetCDF file
- Built on top of MPI-IO which utilizes optimal I/O facilities provided by the parallel file systems
- Allows for MPI-IO collective I/O, datatypes, and hints for further optimization



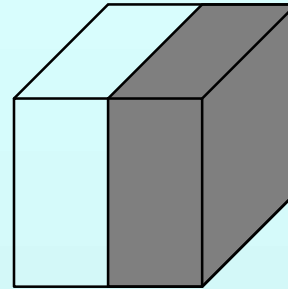
LBNL Benchmark



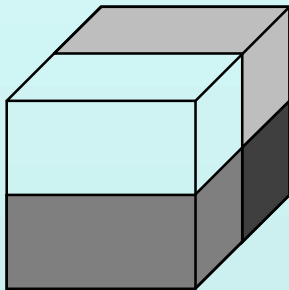
Z partition



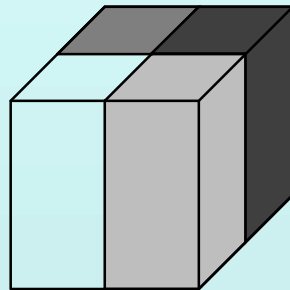
Y partition



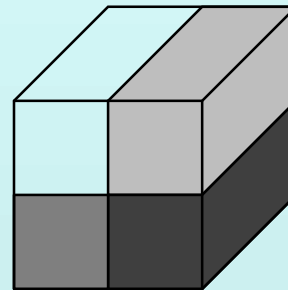
X partition



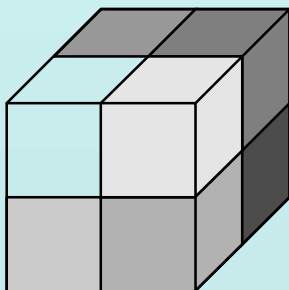
YZ partition



XZ partition



XY partition



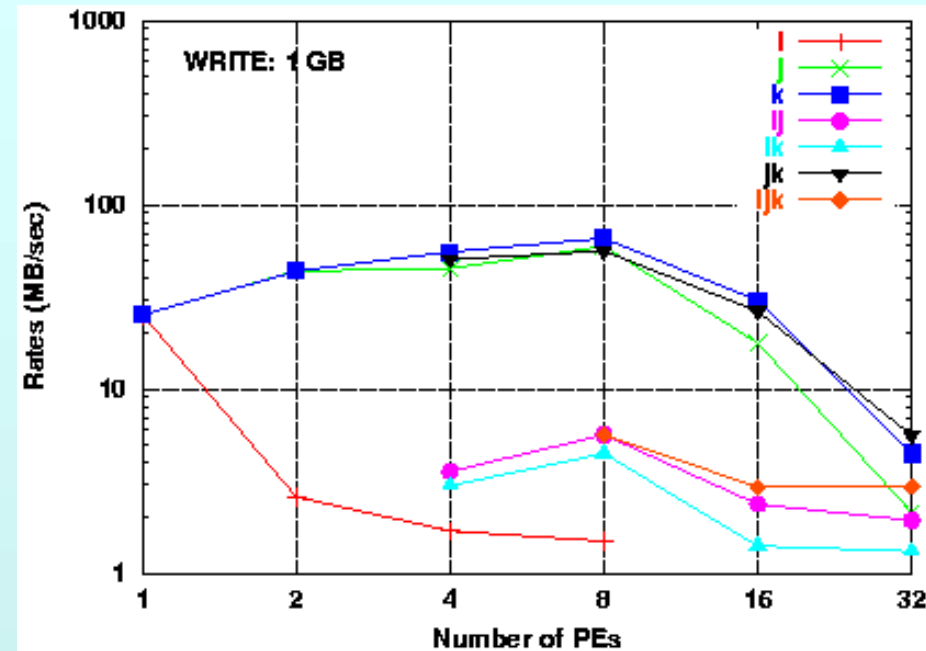
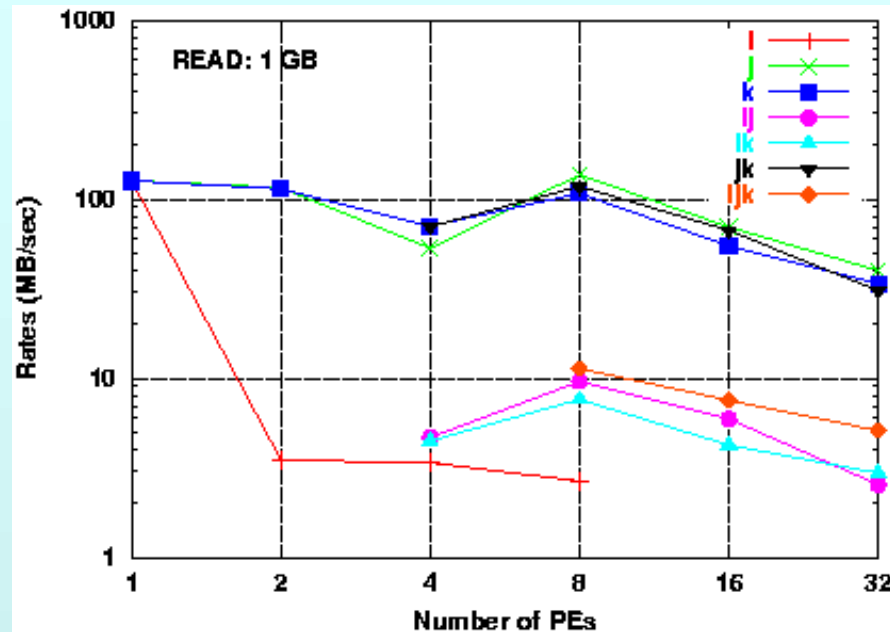
XYZ partition

- | | |
|-------------|-------------|
| processor 0 | processor 4 |
| processor 1 | processor 5 |
| processor 2 | processor 6 |
| processor 3 | processor 7 |

Test suite

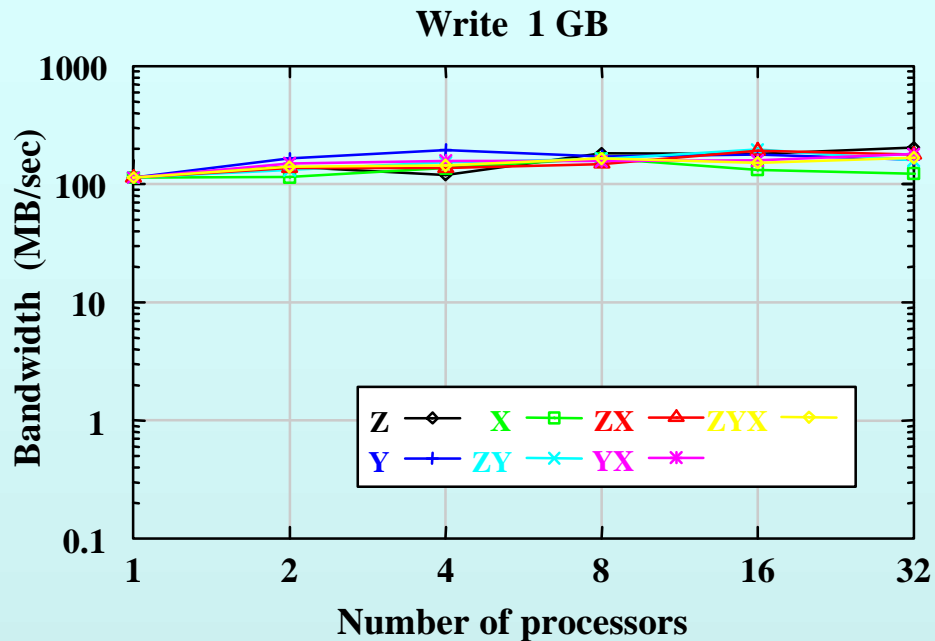
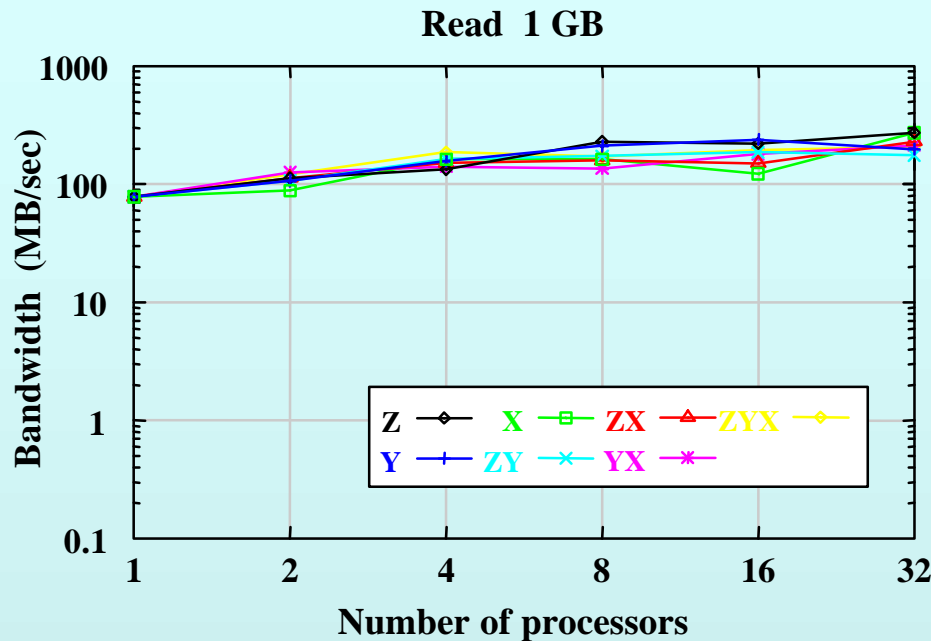
- Developed by Chris Ding et al. at LBNL
 - Written in Fortran
 - Simple block partition patterns
- ## Access to a 3D array which is stored in a single netCDF file
- Running on IBM SP2 at NERSC, LBNL
 - Each compute node is an SMP with 16 processors
 - I/O is performed using all processors

LBNL Results – 1 GB



- ❑ Array size – 512 x 512 x 512, real*8
- ❑ Read
 - No better performance is observed
- ❑ Write
 - 4-8 processor writes results in 2-3 times higher bandwidth than using a single processor

Our Results – 1 GB

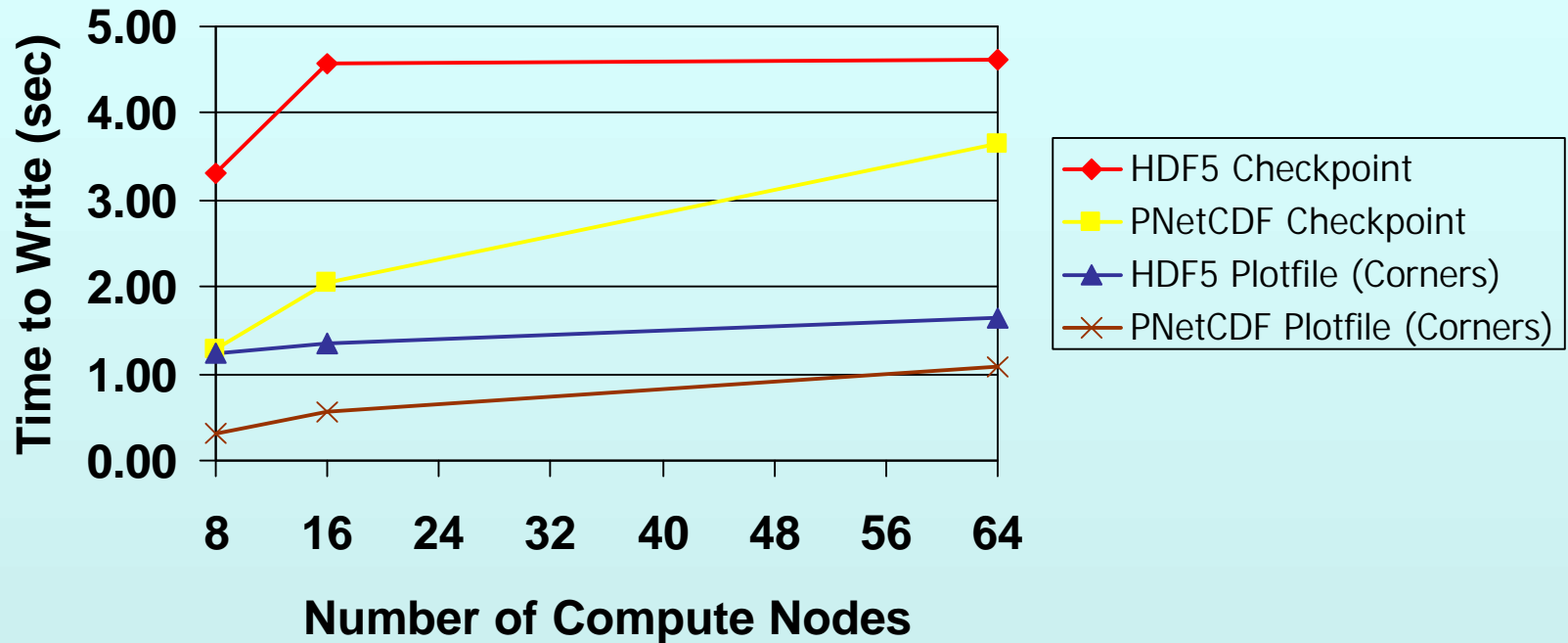


- ❑ Array size: 512 x 512 x 512, real*8
- ❑ Run on IBM SP2 at SDSC
- ❑ I/O is performed using one processor per node

FLASH I/O Benchmark Comparison



Parallel NetCDF vs. HDF5 Performance



- ❑ Ported FLASH I/O Benchmark to Parallel NetCDF
- ❑ Results on IBM SP at SDSC
- ❑ Preliminary numbers - further optimization planned
- ❑ Parallel NetCDF provides a useful subset of HDF5 features that are more amenable to parallel I/O

□ Parallel Virtual File System

- Parallel File System for Linux clusters
- SDM work extending PVFS capabilities to better match scientific application requirements (structured file access)

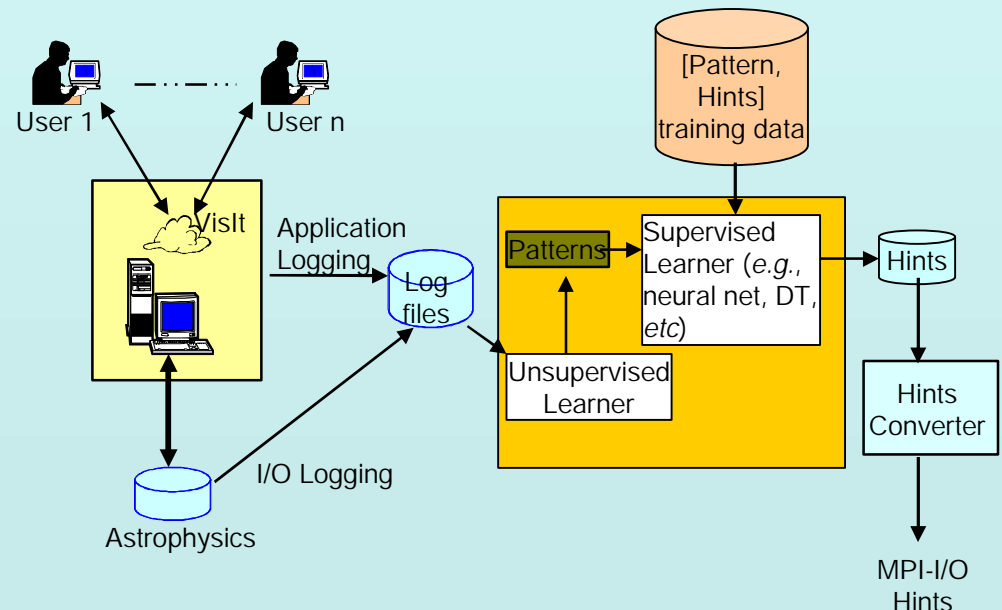
□ ROMIO

- MPI-IO implementation used on most platforms
- SDM work
 - Harnessing new capabilities in PVFS
 - Enabling access to grid I/O resources via MPI-IO interface
 - Extending hints available for performance tuning

Automated Hint Generation



- ❑ Hints to underlying system software can boost performance by creating new pre-fetching and caching strategies
- ❑ “Right” set of hints varies by user, application, and system
- ❑ Automated generation of hints from I/O log files can lead to discovery of “most desirable” set of hints
- ❑ Feedback from previous hints to I/O system can shorten the discovery process
- ❑ Utilizing and extending MPI-IO hint mechanism is an ideal candidate for
- ❑ this approach



ASPECT: Adaptable Simulation Product Exploration & Control Tool



Application Scientists:

- Tony Mezzacappa, ORNL, Astrophysics
- David Erickson, ORNL, Climate
- John Drake, ORNL, Climate

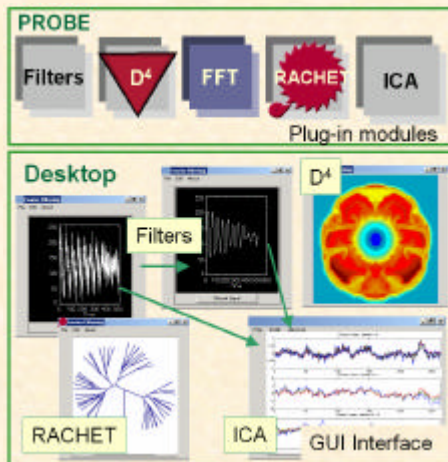
Development Team:

- **Nagiza F. Samatova**
Project Lead
- **George Ostrouchov**
- **Ian Watkins**
- **David Bauer**
- **Guru Kora**
- **Hoony Park**
- **Jennifer Golek**
- **Faisal AbuKhzam**
- **Yongming Qu**

Technology Collaborators:

- Randy Burris, ORNL
- Ross Toedte, ORNL
- Rob Ross, ANL
- Bill Gropp, ANL
- Rajeev Thakur, ANL
- Rob Grossman, UIC
- Alok Choudhary, NWU
- Wei-keng Liao, NWU
- Jim Ahrens, LANL
- Gene Golub, Stanford U.
- Mike Langston, UTK

ASPECT



<http://www.scidac.org/SDM/ASPECT>

Typical Simulation Exploration Scenarios Driven by limitations of existing technologies



❑ Post-processing Scenario:

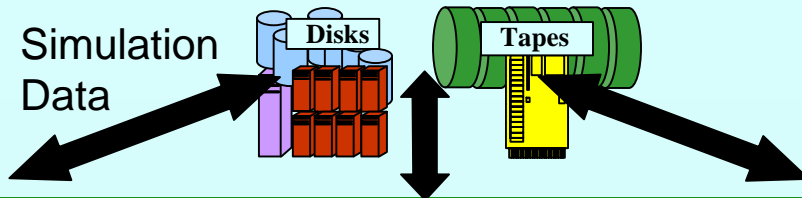
- ⇓ Submit a long-running simulation job (weeks – months)
- ⇓ Periodically check the status (run “tail -f” command on each machine)
- ⇓ Analyze large simulation data set

❑ Real-time Scenario:

- ⇓ Instrument a simulation code to visualize a field(s)
- ⇓ While running a simulation job
 - Monitor the selected field(s)
 - If can not monitor, then either Stop a job or Continue running without monitoring and ability to view later what has been skipped
- ⇓ If changing a set of fields to monitor, then go to 1

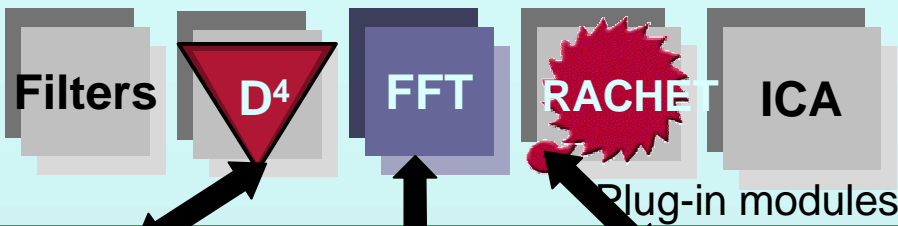
Improvements through — ASPECT

Data stream - not simulation - monitoring tool

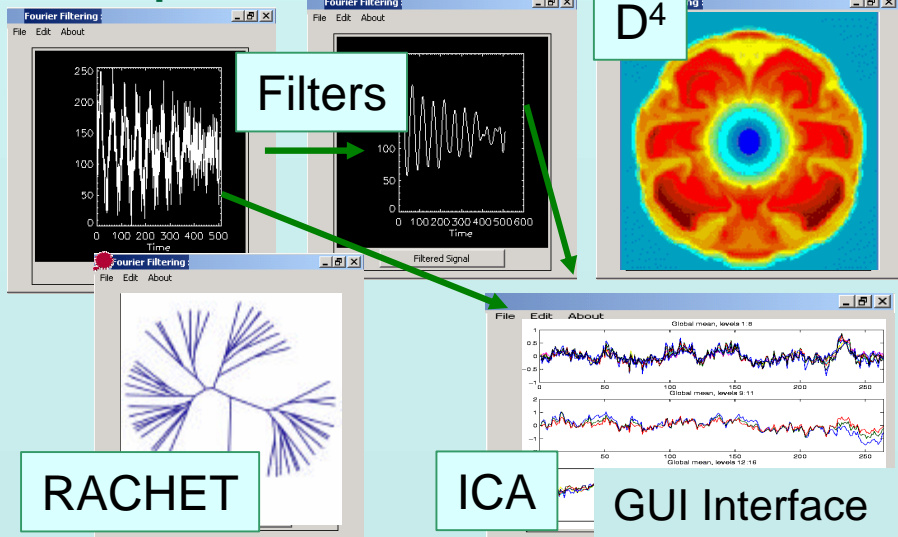


ASPECT

PROBE



Desktop



ASPECT's advantages:

- No simulation code instrumentation
- Single data — multiple views of data
- No interference w/ simulation
- Decoupled from the simulation

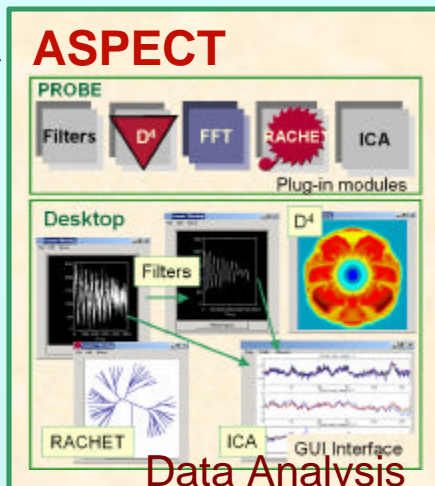
ASPECT's drawbacks:

- No computational steering
- No collaborative visualization

“Run and Render” Simulation Cycle

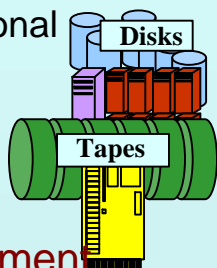
ASPECT Features:

- ✓ Enables effective and efficient monitoring of data generated by long running simulations.
- ✓ Provides the GUI interface to a rich set of pluggable data analyses.
- ✓ Supports scientific & statistical data analysis visualization via pVTK.
- ✓ Handles large data sets via data reduction & parallel algorithms
- ✓ Provides efficient I/O through MPI-IO to NetCDF and HDF.
- ✓ Transfers data efficiently through UDB-based Sabul protocol.



PROBE for Storage & Analysis of Simulation Data:

- High-Dimensional
- Distributed
- Dynamic
- Massive



Data Management

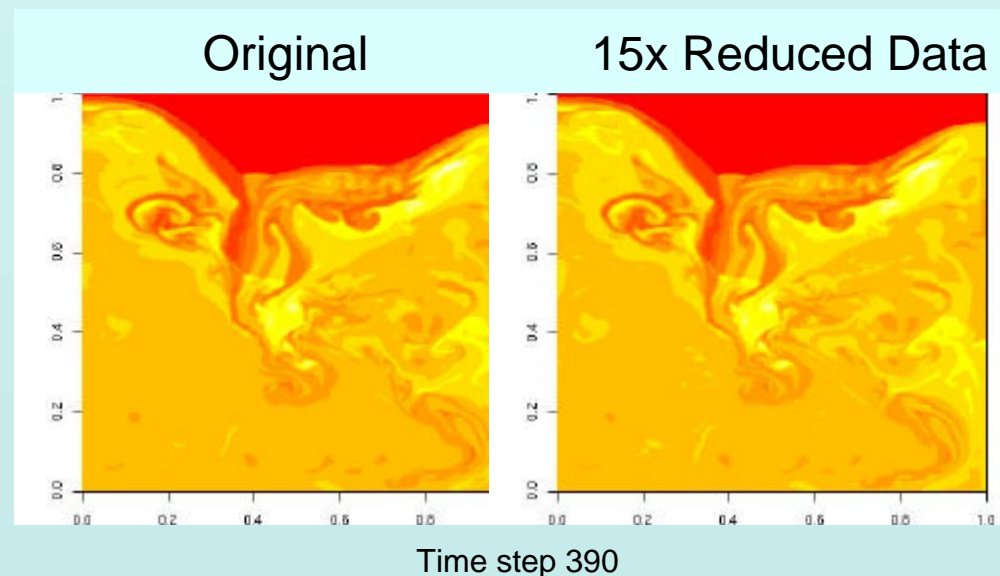
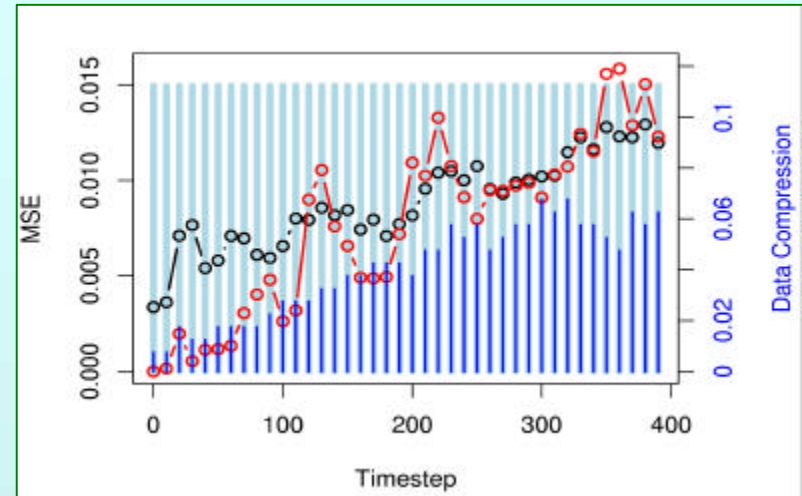
Terascale
Supernova
Explosion
(TSI) Simulation
Computational
Environment



Application Scientist

Adaptive Data Reduction in ASPECT

- ❑ **SciDAC TSI simulation:**
 - 15 to 200 times reduction per time step
 - Outperforms sub-sampling 3 times for comparable MSE over all time steps
 - Provides 30-fold compression with 99% accuracy (captured variability)
- ❑ Based on SVD of contiguous field blocks
- ❑ Exploits spatial correlation & adapts to complexity of spatial field
- ❑ Parameter controls captured percentage of variation
- ❑ Linear time field restoration

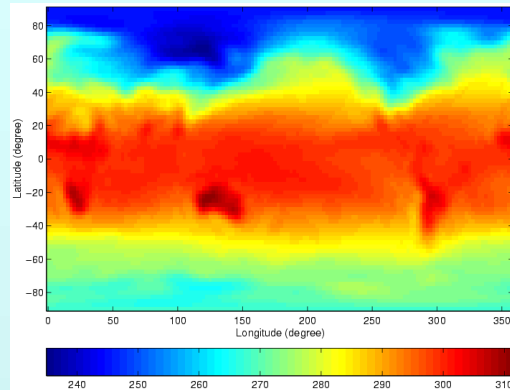


This work is expanding the scientific understanding of global climate change



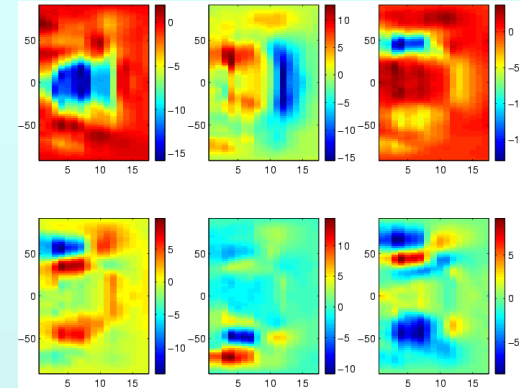
- ❑ **Goal:** understand changes in global temperatures
- ❑ **Problem:** separation of different sources, such as El Niño (ENSO) and volcano eruptions
- ❑ **Results:** first to identify ENSO component in zonally averaged re-analysis climate data from NCEP
- ❑ **Next steps:** refine techniques and scale up algorithms to identify more sources in other datasets
- ❑ **Collaborations:** vital to our success
 - **Dr. B. D. Santer:** climate expert
 - **SciDAC team:** latest computational methods
- ❑ **Other benefits:** general purpose software tools for dimension reduction and source separation; re-usable in other domains; synergistic to efforts at ORNL and LBNL
- ❑ **Work performed by:** Imola Fodor, Chandrika Kamath (LLNL)

New Algorithm Successfully Isolates the ENSO Component in Global Temperatures



Dimension reduction

PCA



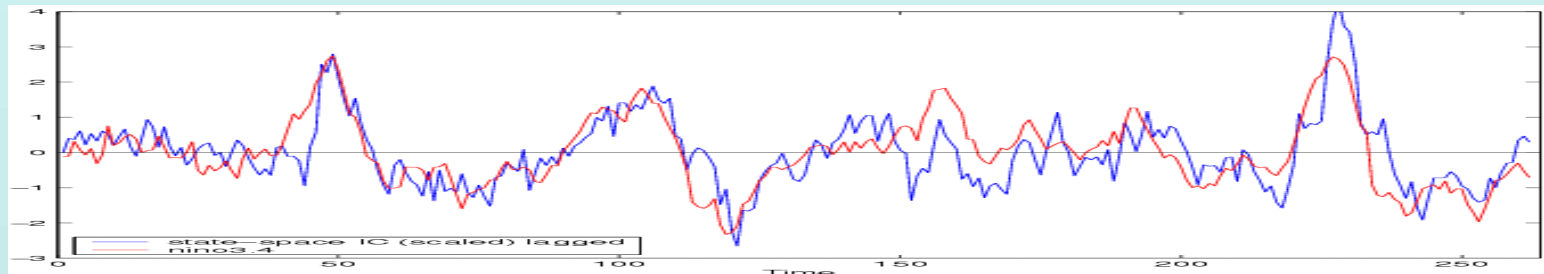
PC basis: 22x73x17

Source separation

reduced dimension

ICA

Raw data: 264x144x73x17
time lat lon altitude



Estimated ENSO source component and Nino 3.4 ENSO index: 264x1
The excellent match indicates the success of our approach.

Distributed, Heterogeneous Data Integration

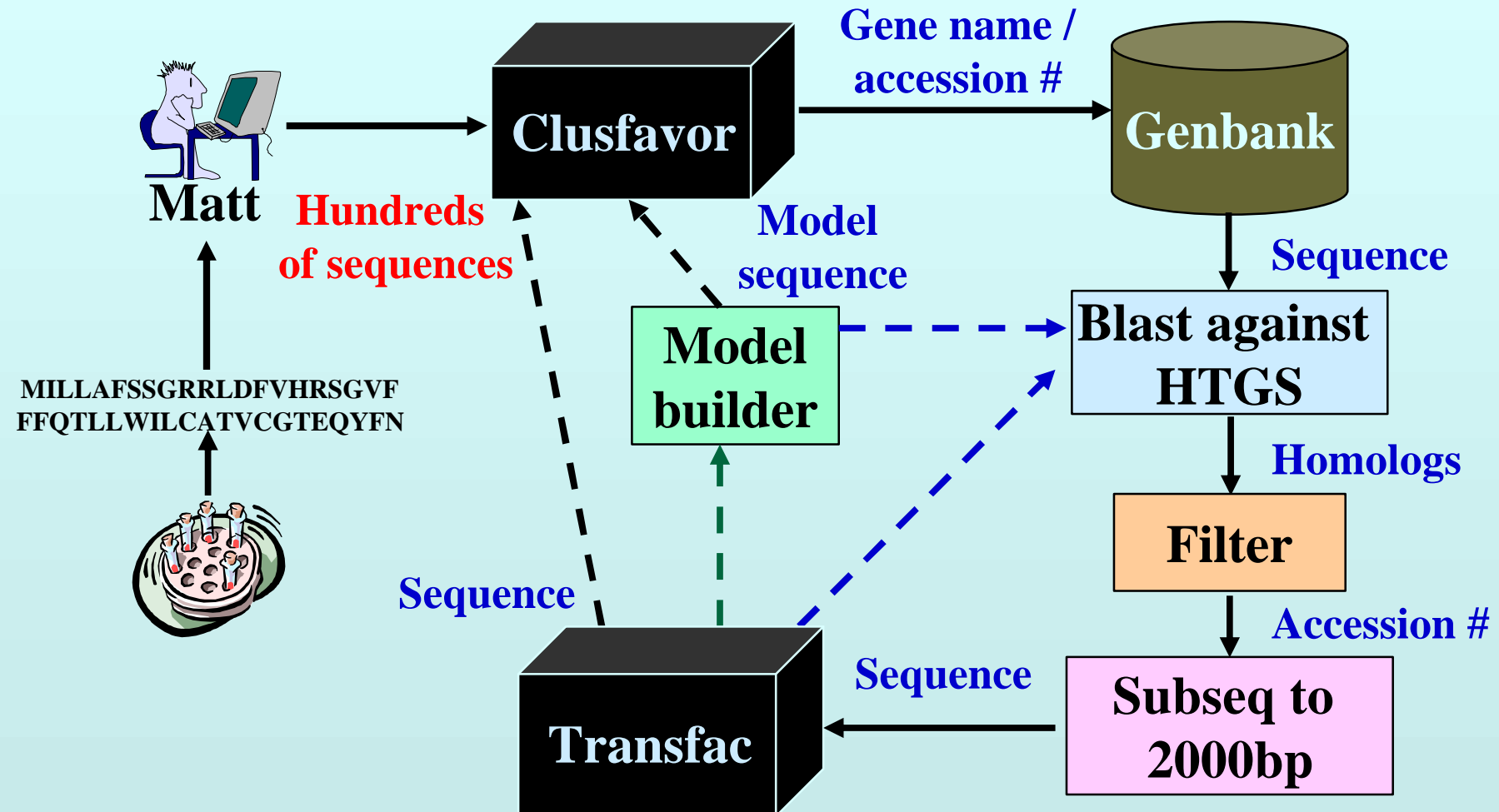


□ Team members

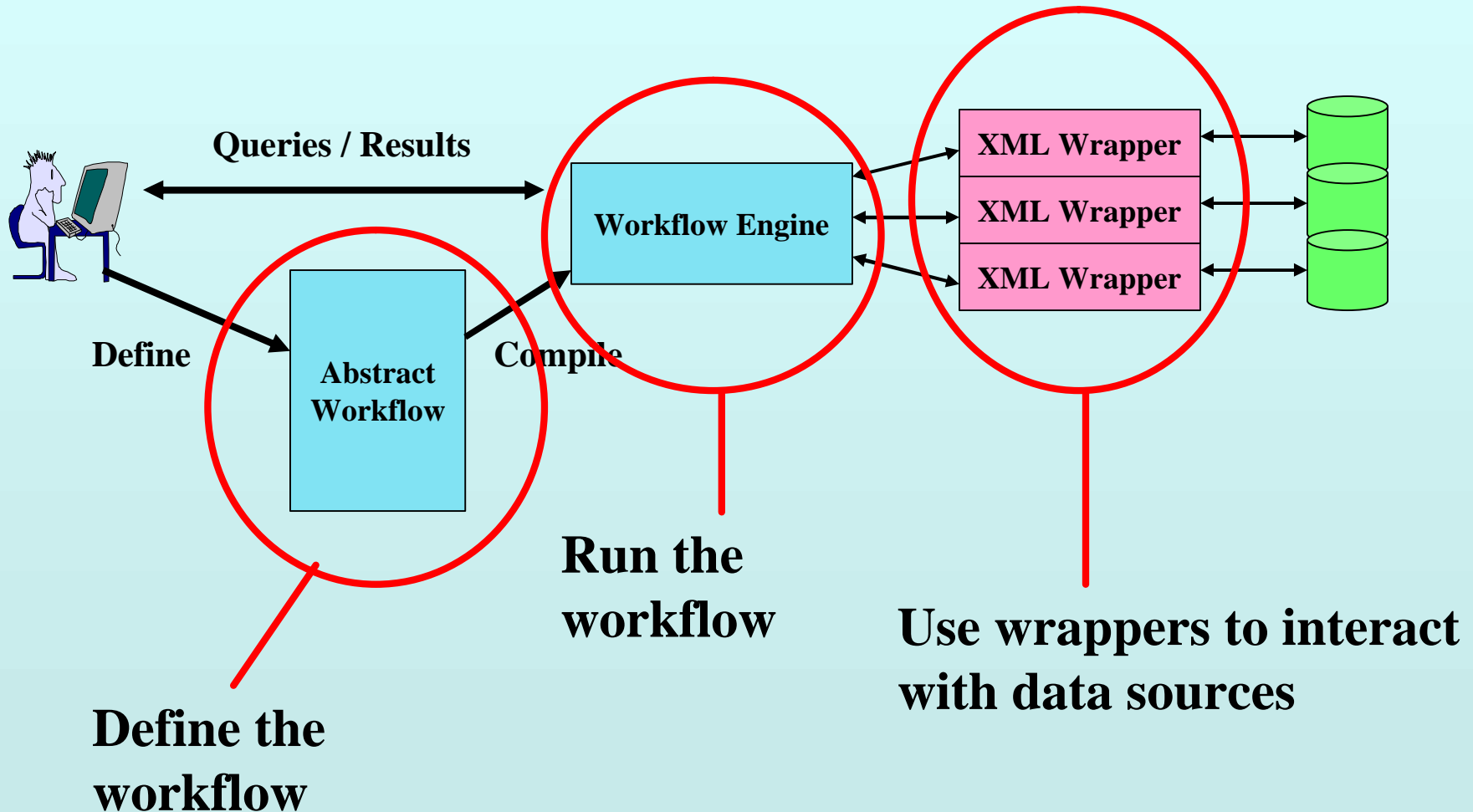
- Terence Critchlow (LLNL)
- Calton Pu, Ling Liu (Georgia Tech)
- Bertram Ludaescher, Amarnath Gupta, Ilkay Altintas (SDSC)
- Mladen Vouk, Donald Bitzer, Munindar Singh, David Rosnick (NCSU)
- **Matt Coleman (LLNL)**

□ Helping scientists perform the complex data manipulations they need to perform their research

Motivating Use Case: Identifying Model Sequences



Workflow based approach to data integration



Query 4

Blast Detail Wrapper

Query 3

Blast Sum Wrapper

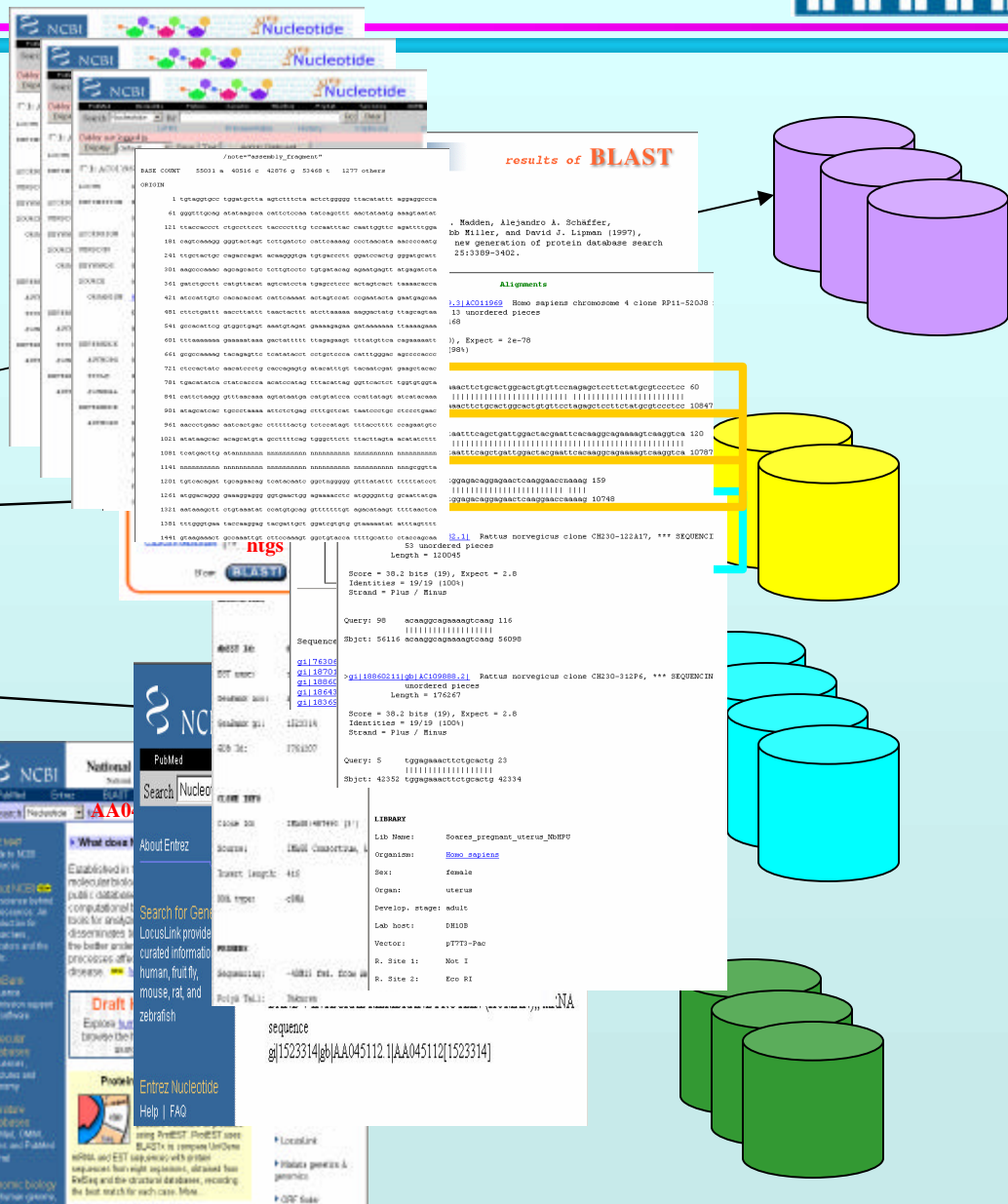
Query 2

Sequence Wrapper

Query 1

Seq. Link Wrapper

Extracting Data from a single Web Document

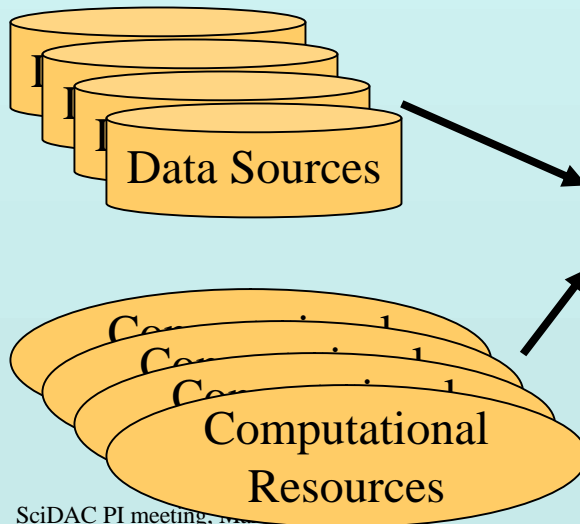


Architecture



Complex Workflow Execution

An extended version of an open source workflow engine executes the workflow.



GUI

Context Mediation

Semantic Mediation

Workflow Support

Registries

Information Wrappers

Web Services

Grid Middleware &
Infrastructure resources

Abstract Workflow Definition

Domain specific transformations use semantic mediation to map the abstract workflow into an executable format.

Automatic Wrapper Generation

The workflow engine is isolated from the data sources by wrappers.

❑ Semantic mediation

- Define abstract workflow language
- Compile abstract workflow to executable workflow
 - Targeting XPDL (workflow engine input format)

❑ Extend open source workflow engine

- Integrate open source workflow engine with existing workflow design tool

❑ Wrapper generation

- Automatically define WSDL and SOAP interfaces

A prototype is being used by Matt Coleman at LLNL

FastBit: Compressed BitMap Index



Applied to:
HENP
Combustion

John Wu, Arie Shoshani, Doron Rotem
LBL

HENP Collaborators: Jerome Lauret, STAR-BNL
Wei-Ming Zhang, Kent State University

Combustion Collaborators: Wendy Koegler,
Jackie Chen, SNL

- ❑ **Need to search over**
 - millions of objects (100s million events)
 - Hundreds of searchable attributes
- ❑ **Most users specify range conditions on a handful of attributes**
- ❑ **Our method (FastBit) is effective with high cardinality attributes by applying:**
 - Binning the attribute values
 - effective compression of bitmaps
 - Optimize Compression for Computation Efficiency

Basic Bitmap Index

- Bitmap index idea: have one bitmap for each value
- For scientific data: values are numeric and have large cardinality
- partition into bins (e.g. 100): 0-10, 11-20, ...

Number of Pions

1	0	0	0	0	0	0
0	1	0	0	0	0	0
0	0	0	0	0	0	1
0	0	0	1	0	0	0
0	1	0	0	0	0	0
0	0	1	0	0	0	0
0	0	1	0	0	0	0
0	0	1	0	0	0	0
0	1	0	0	0	0	0
0	0	1	0	0	0	0
0	0	1	0	0	0	0
0	0	1	0	0	0	0
0	0	1	0	0	0	0
0	0	0	0	0	1	0
0	0	1	0	0	0	0
0	0	1	0	0	0	0

0	1	0	0	0
0	0	1	0	0
0	1	0	0	0
0	1	0	0	0
0	0	0	0	0
1	0	0	0	0
1	0	0	0	0
1	0	0	0	0
0	0	1	0	0
1	0	0	0	0
1	0	0	0	0
1	0	0	0	0
1	0	0	0	0
0	0	0	1	0
1	0	0	0	0
1	0	0	0	0

The basic
bitmap index

...

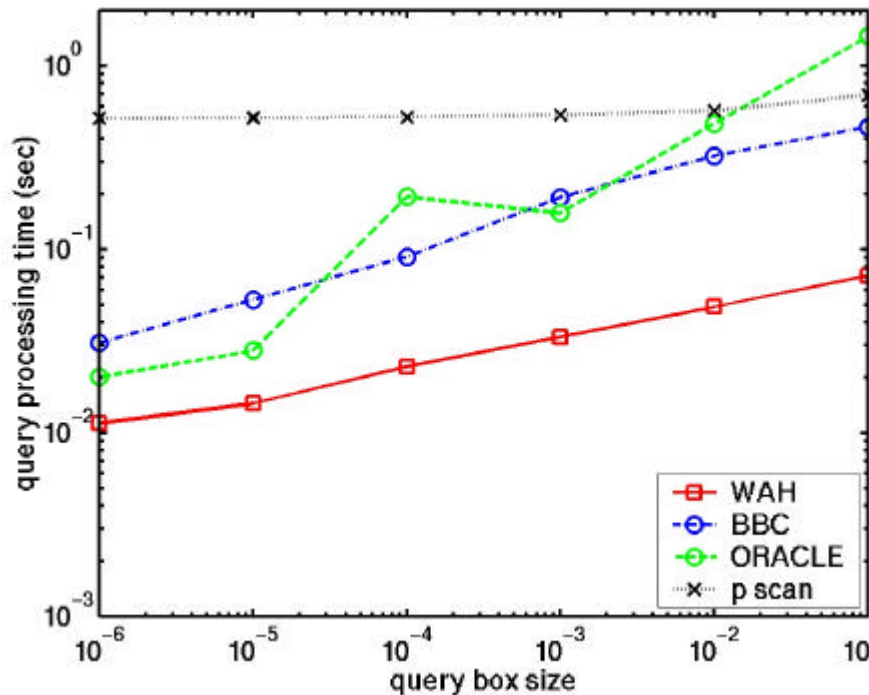
Energy

0	0	0	0	1	0	0	0
0	0	0	0	1	0	0	0
0	0	0	0	1	0	0	0
0	0	0	0	1	0	0	0
1	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0
0	0	0	1	0	0	0	0
0	0	0	1	0	0	0	0
0	0	0	1	0	0	0	0
0	0	1	0	0	0	0	0
0	0	0	1	0	0	0	0
0	0	0	1	0	0	0	0
0	0	0	1	0	0	0	0
0	0	0	0	0	0	1	0
0	0	0	1	0	0	0	0
0	0	0	1	0	0	0	0

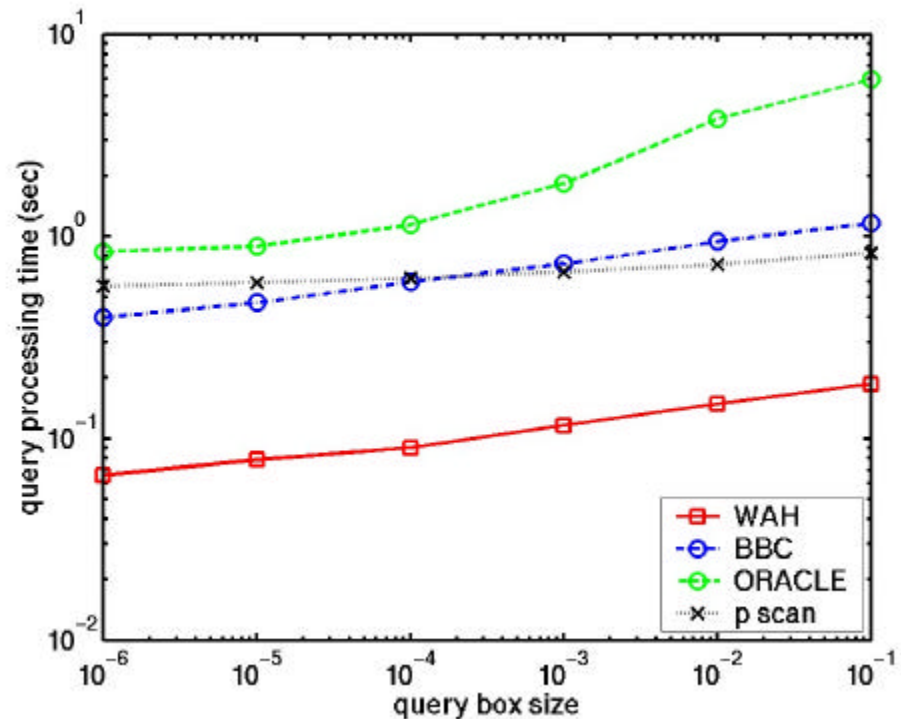
- Only a single “1” in each row
- WAH: Optimize Compression for Computation Efficiency

Performance of Partial Range Queries

2.2 million records of STAR data



2 attributes per query



5 attributes per query

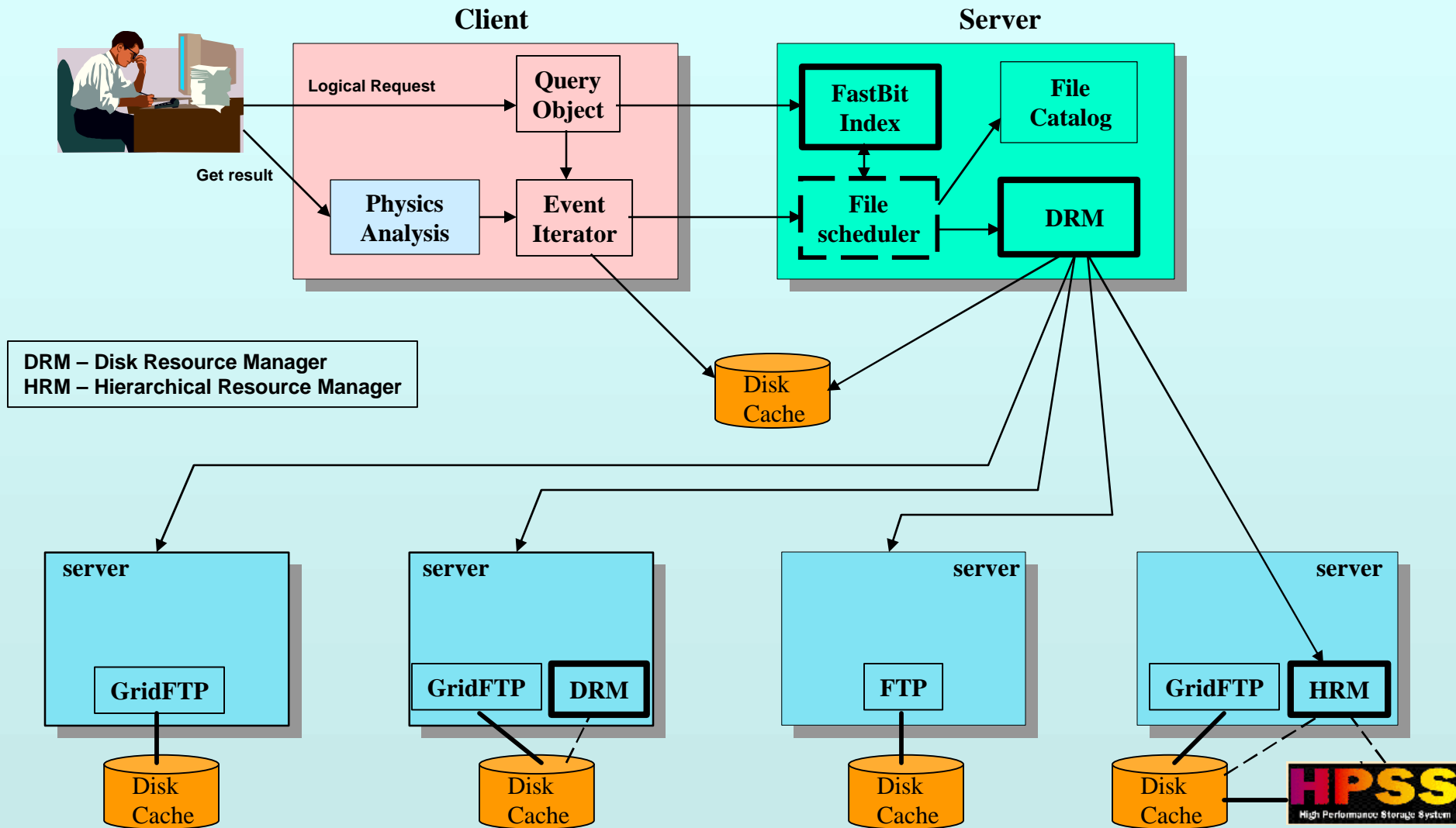
- WAH compressed indexes are
 - >10X faster than ORACLE,
 - >5X faster than our implementation of BBC (BBC is also used in ORACLE)
- 12 most queried attributes are used, average attribute cardinality 222,000

Using Bitmap Indexing in the STAR Analysis Framework



- ❑ **Generate large amounts of raw data**
 - Collect from experiments or large simulations
 - STAR: ~ 100 Million collision events a year
 - STAR: Each event ~1-5 MB
- ❑ **Post-processing of data**
 - process data (find particles produced, tracks)
 - generate summary data
 - e.g. momentum, no. of pions, transverse energy
 - Number of properties is large (50-100 attributes)
- ❑ **Analyze data**
 - use summary attributes to select relevant events
 - extract subsets from the large dataset
 - Need to access events based on partial properties specification (range queries)
 - e.g. $((0.1 < \text{Energy} < 0.2) \wedge (10 < N_p < 20)) \vee (N > 6000)$
 - Current practice: generate pre-selected subsets (called micro-DST)

Architecture for Dynamic Analysis Framework



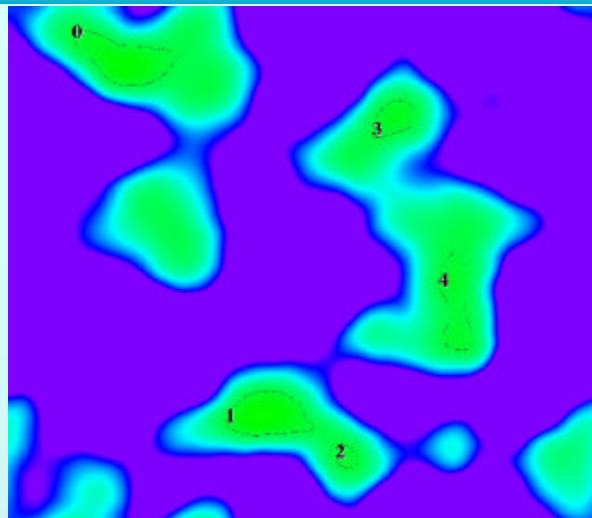
Discovering Combustion Flame Fronts using Bitmap Indexing Technology



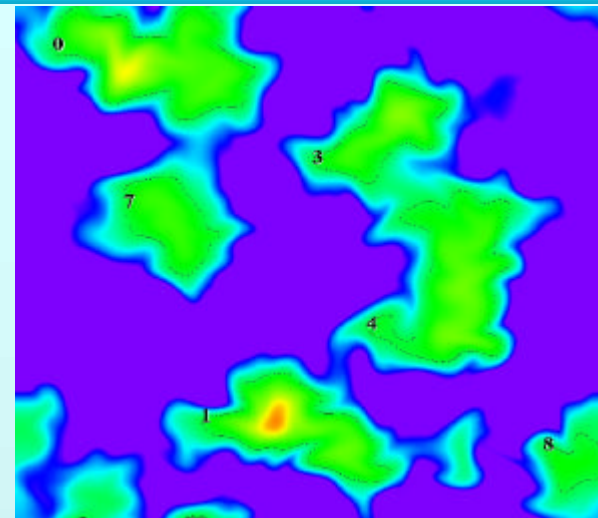
- ❑ **Characteristics of feature tracking in combustion data analyses**
 - High-fidelity simulation on 2D or 3D uniform grids, say, 1000 x 1000 x 1000, for hundreds of time steps
 - A model of hydrogen-air mixture has a dozen attributes per grid point, a realistic model has many more
 - Features are defined to be regions that satisfy the user specified conditions, such as,
“600 < Temperature < 700 AND HO₂ concentration > 10⁻⁷”
- ❑ **Goal:**
 - Interactive feature tracking
 - Multiple attribute conditions for specifying regions
- ❑ **We use FastBit for:**
 - Cell identification
 - Region growing
 - Region tracking

Tracking Features Over Time Steps

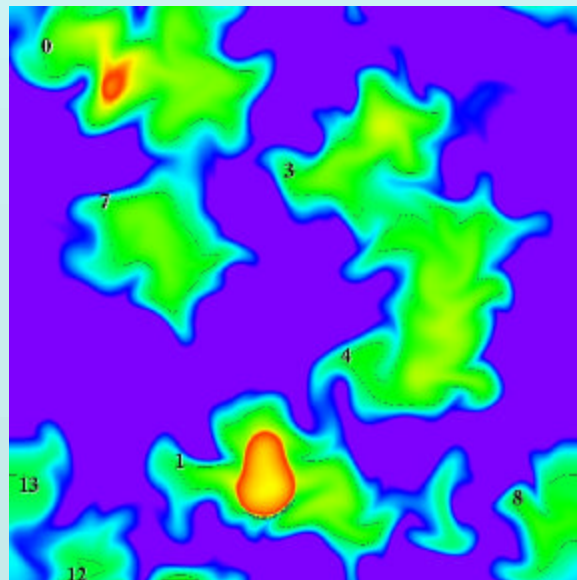
t_1



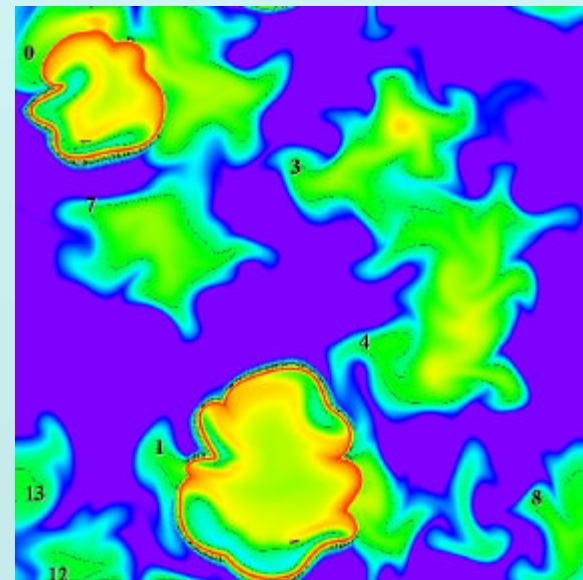
t_2



t_3



t_4



- Features are outlined with thin black lines

- Red color indicates a high concentrate of a transient chemical HO_2

Performance- Feature Tracking



Average time (seconds) to perform steps of feature tracking

600 x 600 grid, 69 time steps (795 MBs)				1344 x 1344 grid, 335 time steps (19.3 GBs)			
# attr	Search	Grow	Track	# attr	Search	Grow	Track
1	1.06	0.22	0.02	1	5.71	2.05	0.12
2	1.67	0.17	0.01	2	7.39	1.24	0.12
3	2.12	0.14	0.01	3	8.92	0.58	0.11
4	2.62	0.14	0.01	4	10.30	0.47	0.10

- Feature tracking in interactive time -- the time is less than ten seconds in most cases
- Feature identification (searching and region growing) in one time step on 1000^3 grid may be completed in less than ten seconds
- Times are for all time steps

Adaptive File Caching and Replication in Distributed Systems



□ Goals

- Develop a coordinated optimal file caching and replication of distributed datasets

□ Two Principal Components of Policy Advisory Module

- A disk cache replacement policy
 - Evaluates which files are to be replaced when space is needed
- Admission policy for file requests
 - Determines which request is to be processed next
 - e.g. may prefer to admit requests for files already in cache

□ Work Performed by:

- Ekow Otoo, Doron Rotem, Arie Shoshani (LBNL)

Some Simulation Results

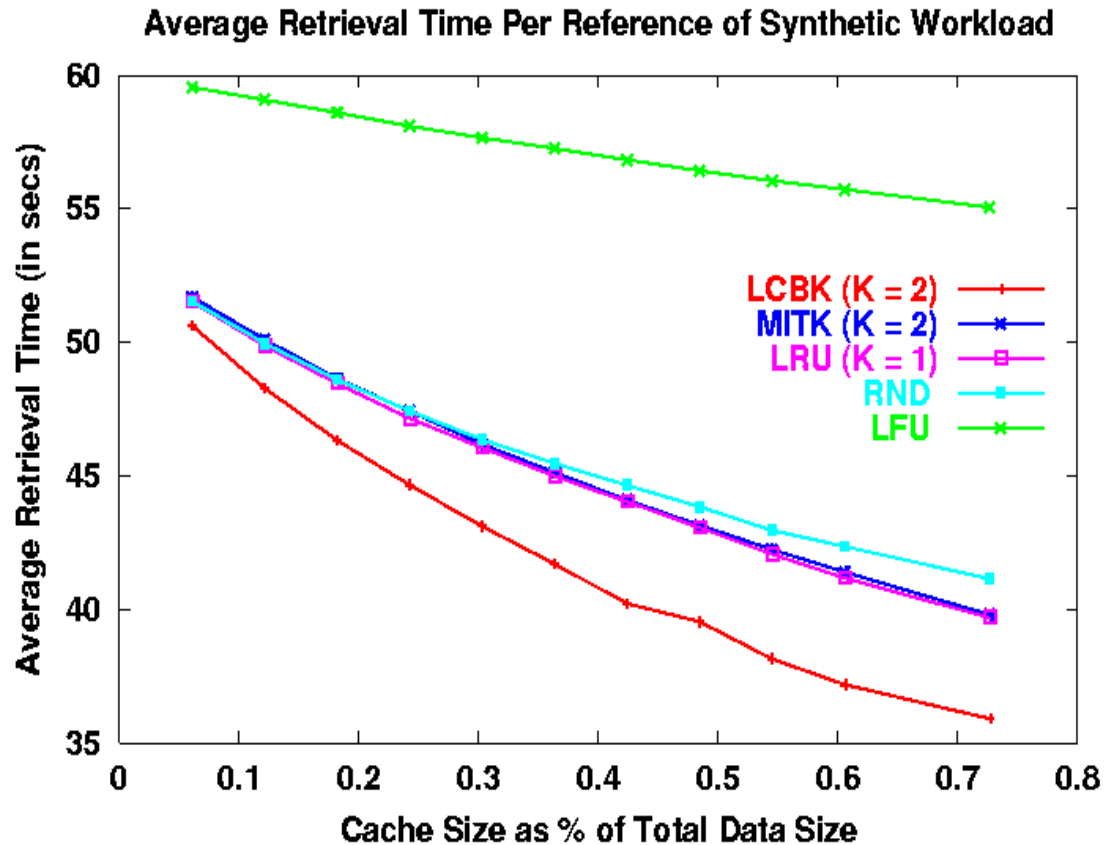


- **LCB-K is the winner**
- **MIT-K, LRU and RND comparable**
- **LFU pretty bad**

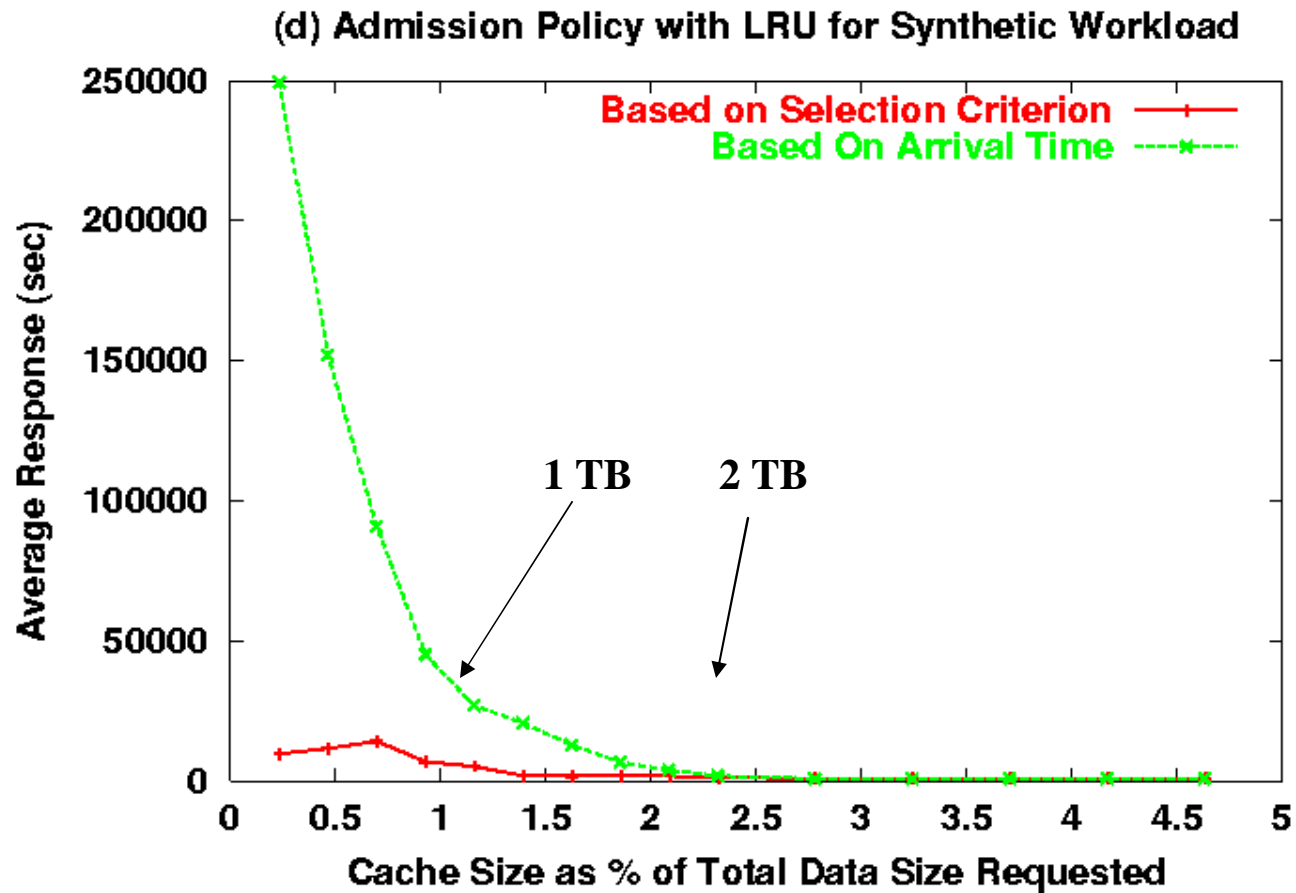
Replacement Policies:

- **RND:** Random
- **LFU:** Least Frequently Used
- **LRU:** Least Recently Used
- **MIT-K:** Maximum Inter-Arrival Time based on last K references
- **LCB-K:** Least Cost Beneficial based on last K references

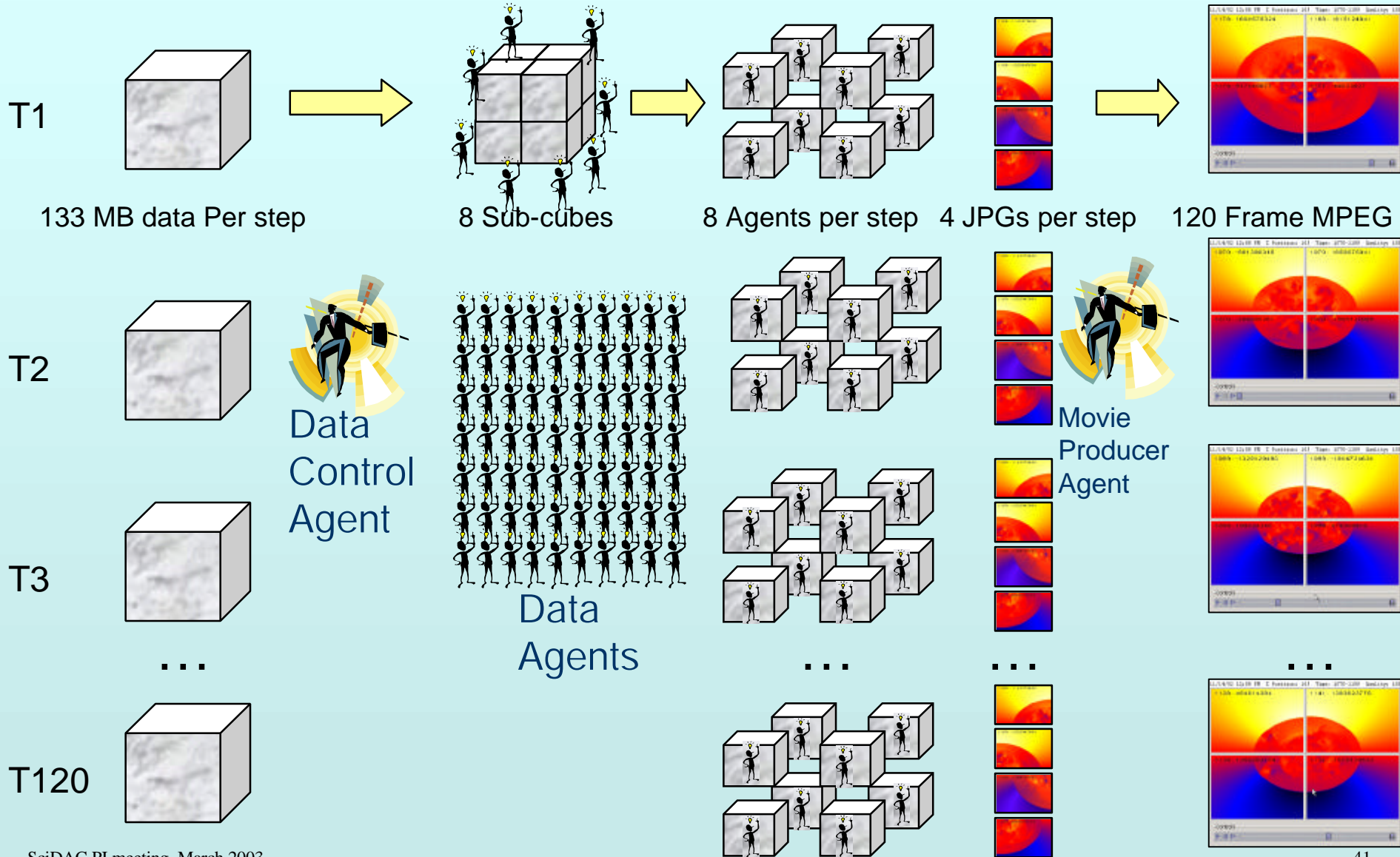
- **Lower values represent better policies**



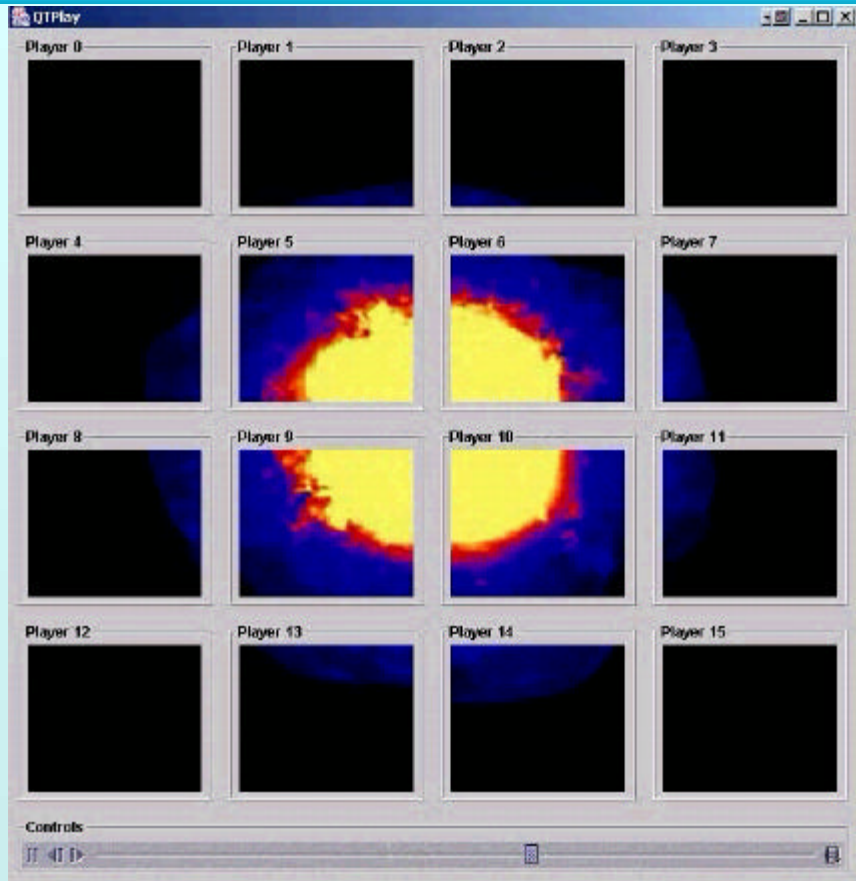
Admission Policy Simulation



A Multi-agent System for Analyzing Massive Scientific Data



Results:

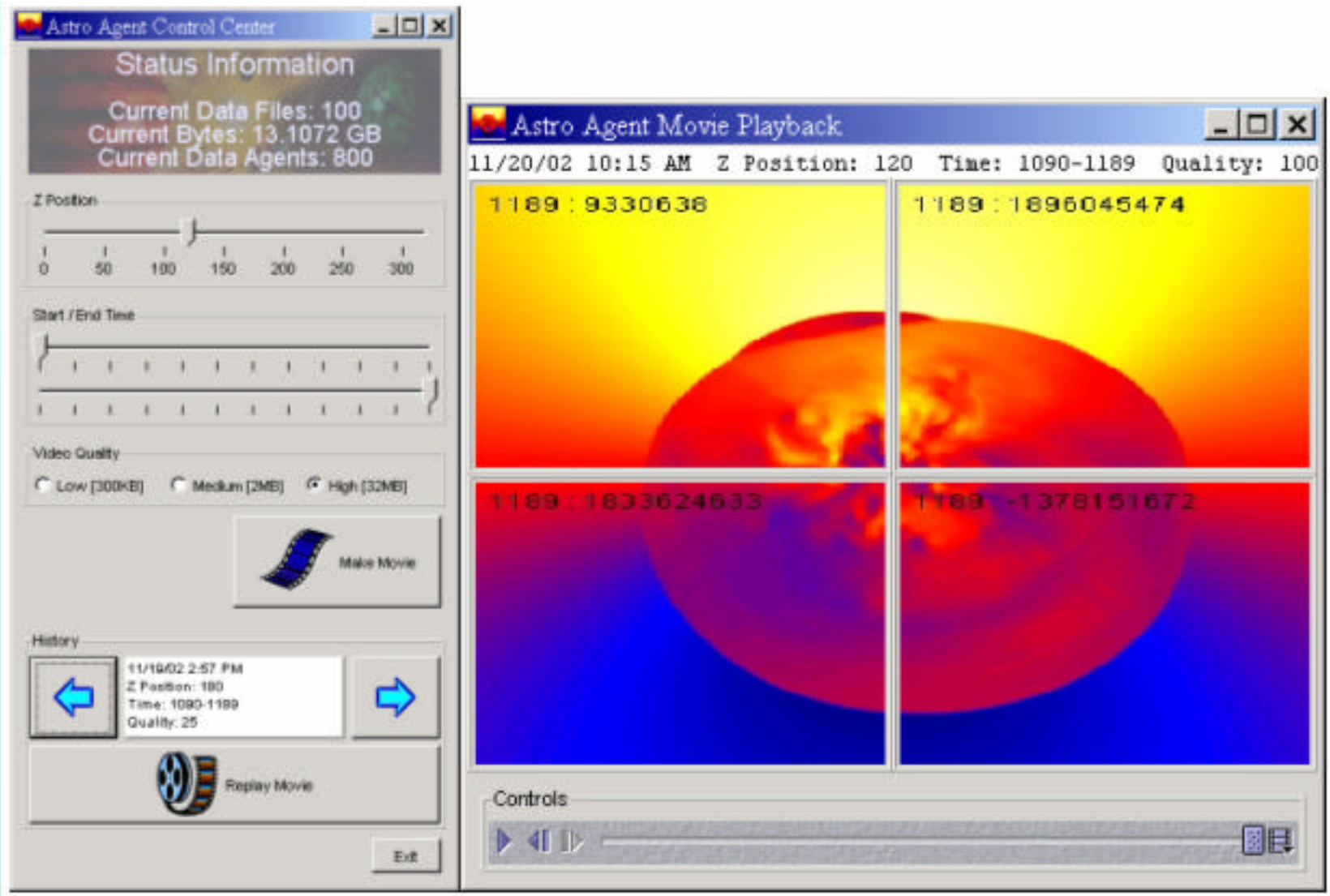


- ❑ Astrophysics data from 194 files
- ❑ 800 agents, each deployed to create an operational picture for 100 time steps
- ❑ Running on three machines
- ❑ Agents automatically created every time a new simulation output file is created

Work performed by:

Thomas Potok, Joel Reed, Tony Mezzacappa, John Blondin, Rick Sheldon

Controlling the movie



Work with individual application scientists



□ Close collaboration with individuals

- Matt Coleman - LLNL (Biology)
- Tony Mezzacappa – ORNL (Astrophysics)
- Ben Santer – LLNL
- John Drake - ORNL (Climate)
- Doug Olson - LBNL, Wei-Ming Zhang – Kent (HENP)
- Wendy Koegler, Jacqueline Chen – Sandia L.
(Combustion)
- Mike Papka - ANL (Astrophysics Vis)
- Mike Zingale – U of Chicago (Astrophysics)
- John Michalakes – NCAR (Climate)

Re-apply techniques to new applications



- ❑ **Parallel NetCDF**
 - Astrophysics → Climate
- ❑ **Adaptive data reduction**
 - Astrophysics → Climate
- ❑ **Compressed bitmaps**
 - HENP → Combustion
- ❑ **Robust File Replication**
 - HENP → Climate
- ❑ **Signal Separation**
 - Climate → Fusion

Summary



- ❑ Our guiding principles served us well
- ❑ We are focused, result oriented
- ❑ Technology migration to new applications
- ❑ Clear future path, lots to do
- ❑ Our focus: getting technology into the hands of scientists